



University of Florida
Sweet Corn and Potato Breeding and Genomics Lab
CANVAS – Salt Lake City, 2025

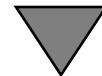


Enhancing Breeding Efficiency: Prediction and Optimization of Breeding Crosses for Long-term Genetic Gain and Diversity

MATE
ALLOCATION
of breeding crosses

Marco Antonio Peixoto* and Marcio Resende

- **Genomic selection:** the use of molecular markers and historical phenotypes to predict new unseen individuals.
- It has **revolutionized** animal/plant breeding programs.
 1. Selection accuracy
 2. Shorten breeding cycles



High performance

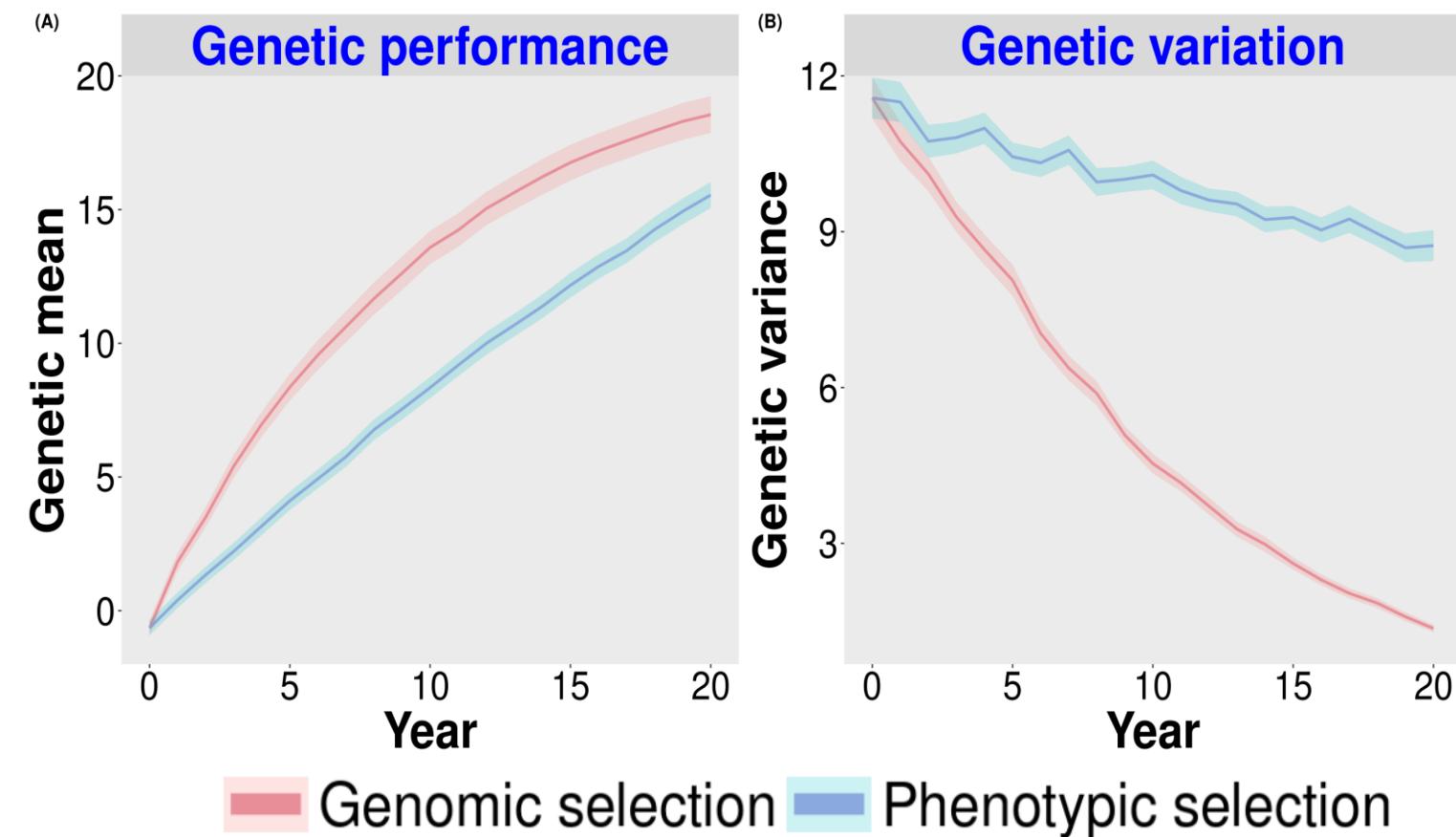
Rationale

- **Genomic selection:** the use of molecular markers and historical phenotypes to predict new unseen individuals.
- It has **revolutionized** animal/plant breeding programs.

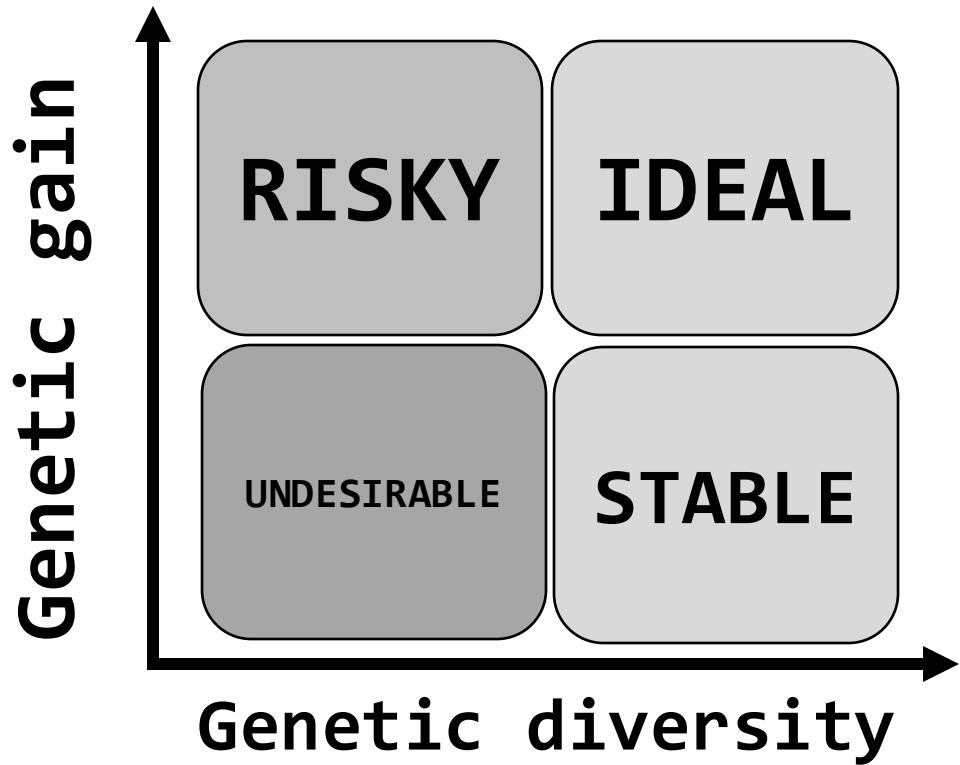
1. Selection accuracy
2. Shorten breeding cycles



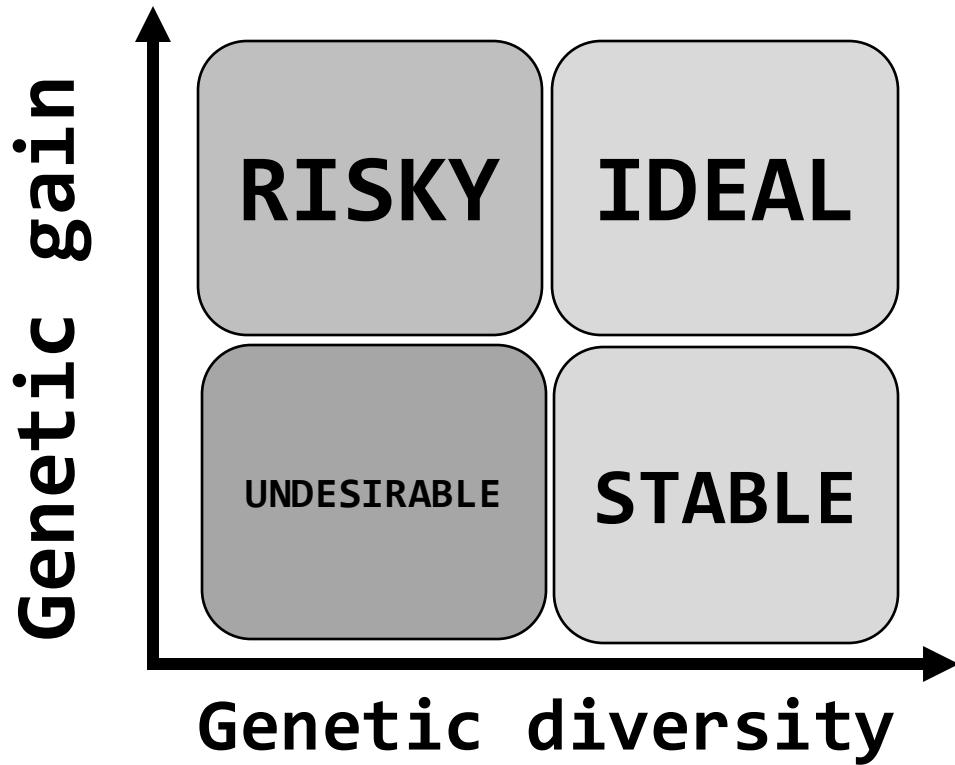
High performance



Problem



Problem

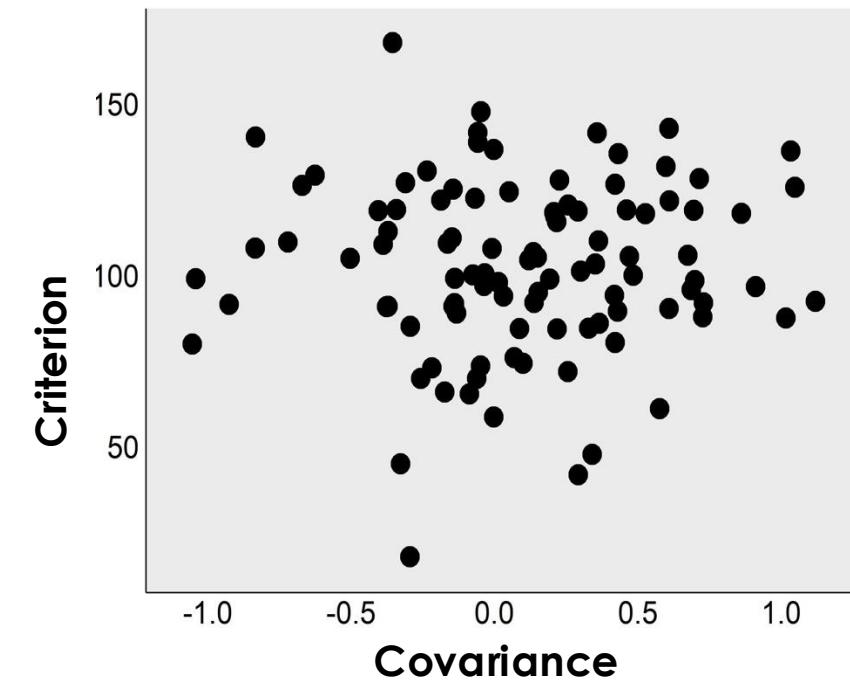
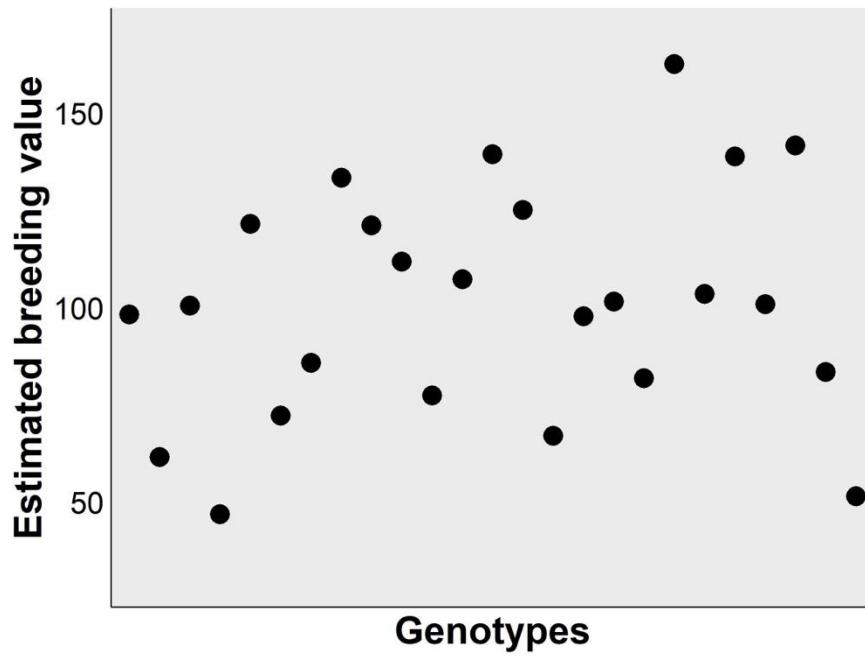


How do we balance rapid improvement with long-term sustainability in breeding programs?

Selection of parental combinations: optimal cross selection

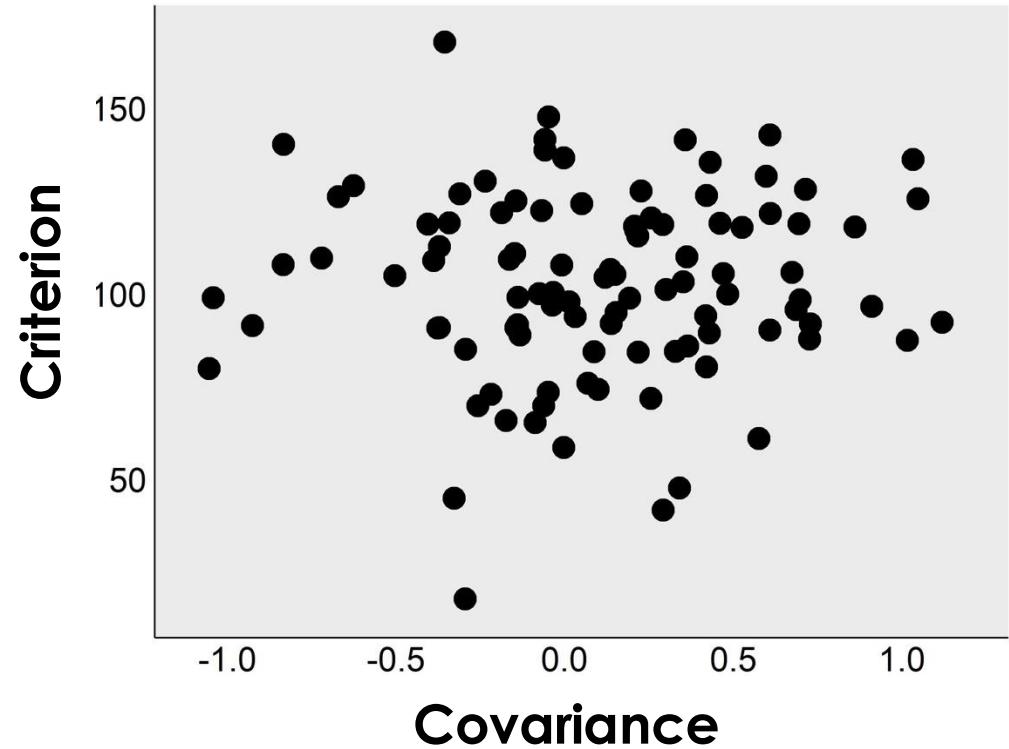


Selection of parental combinations: optimal cross selection

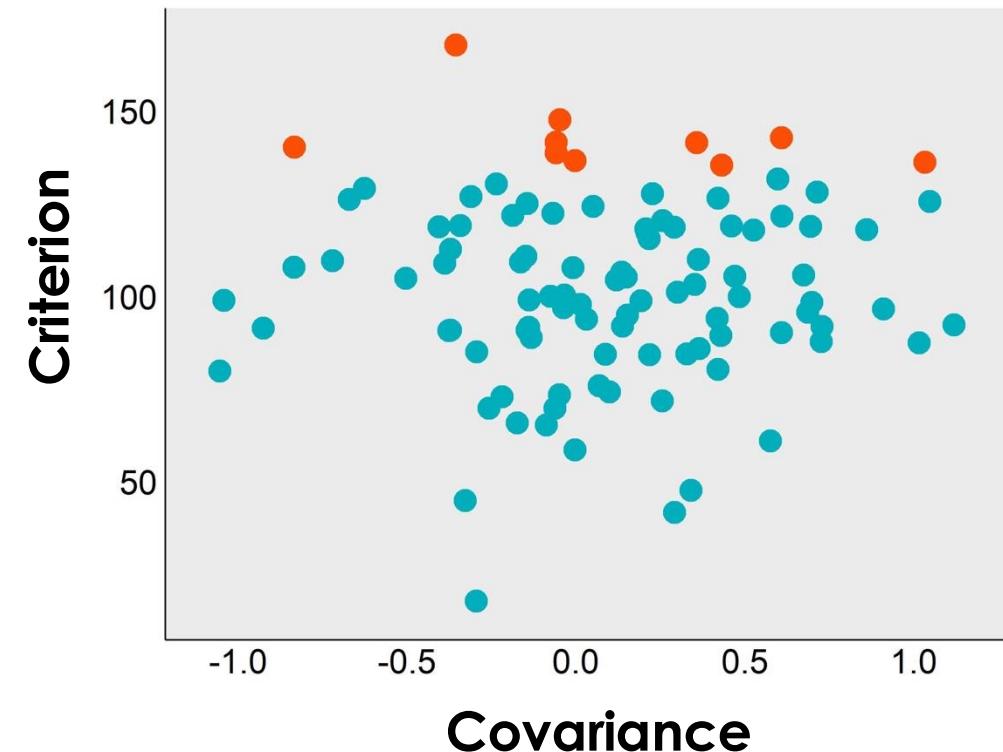
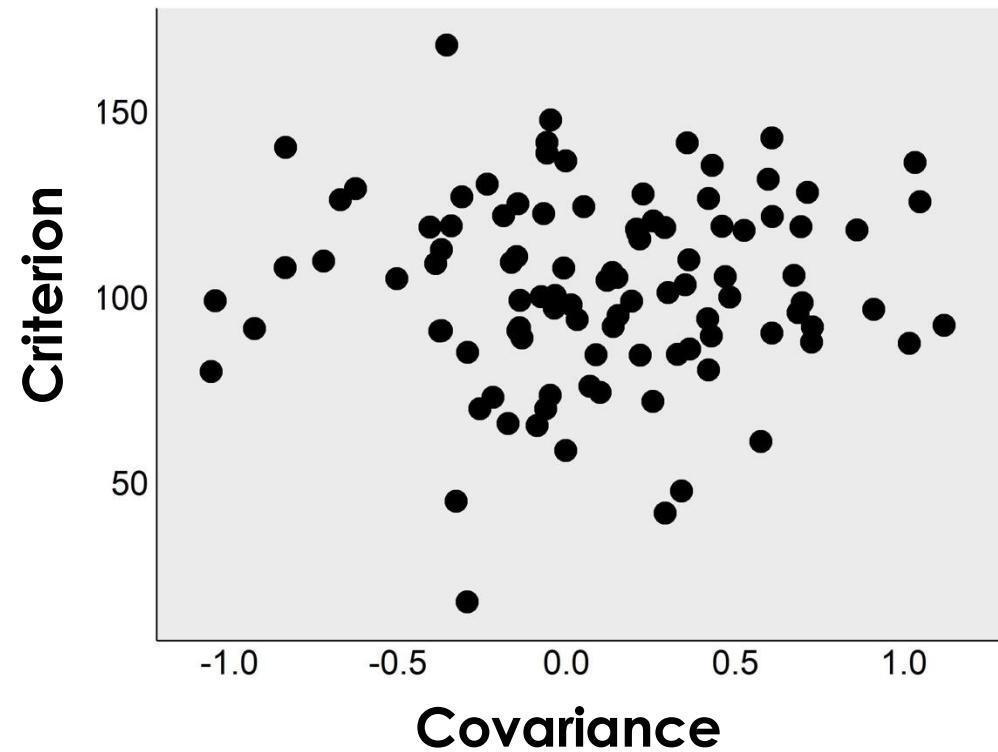


$$\mu_{ebv} = \frac{EBV_{p1} + EBV_{p2}}{2}$$

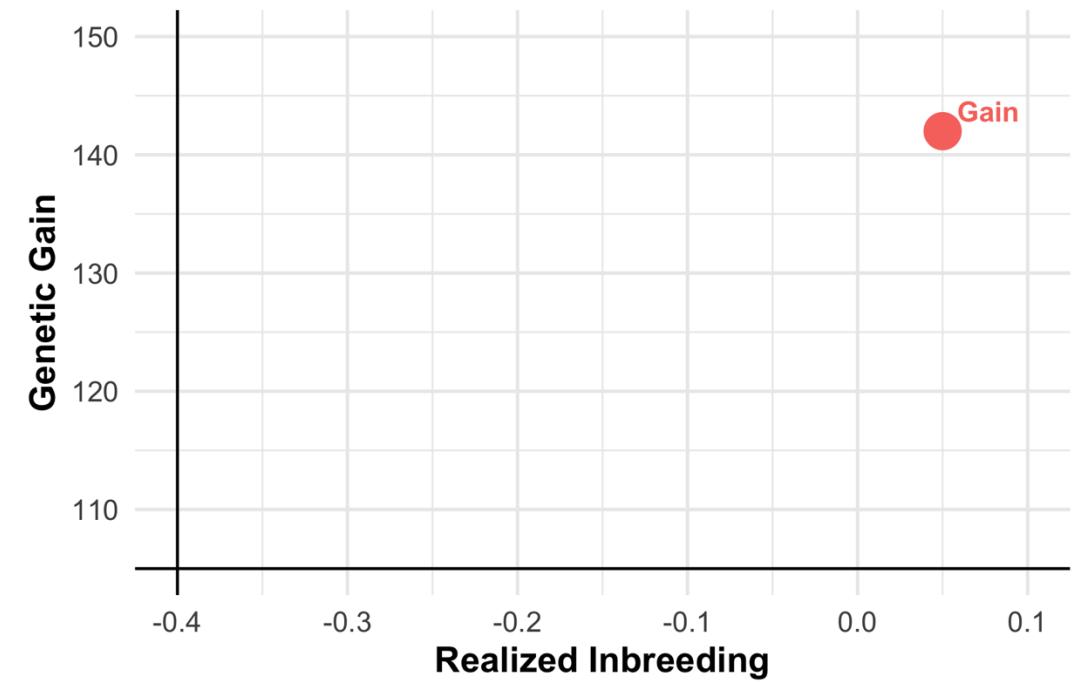
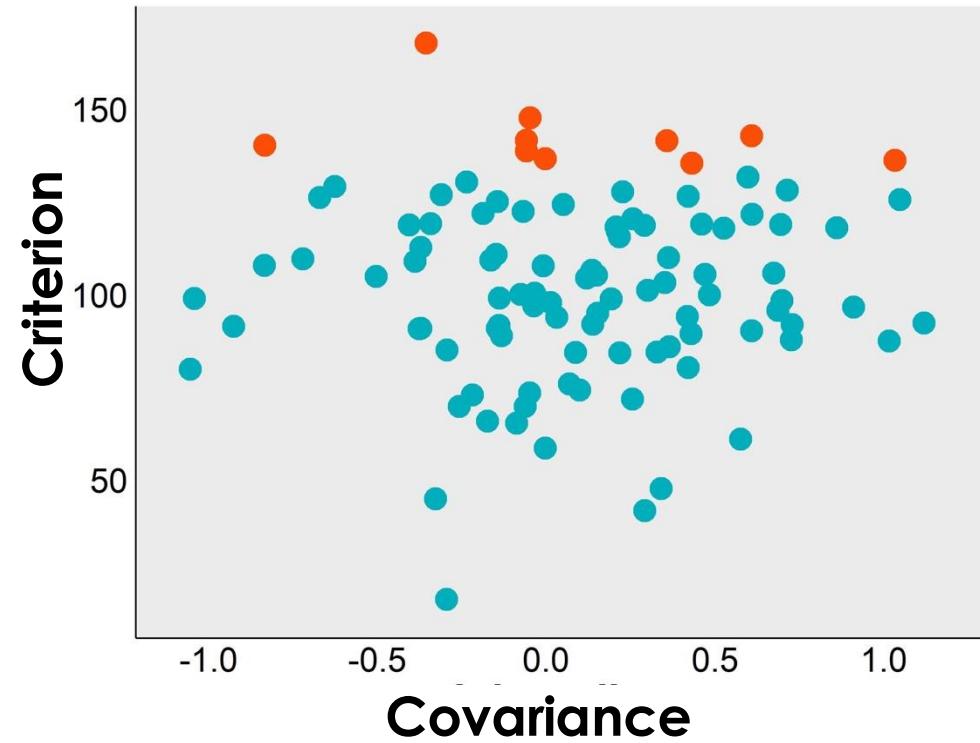
Optimal cross selection



Optimal cross selection



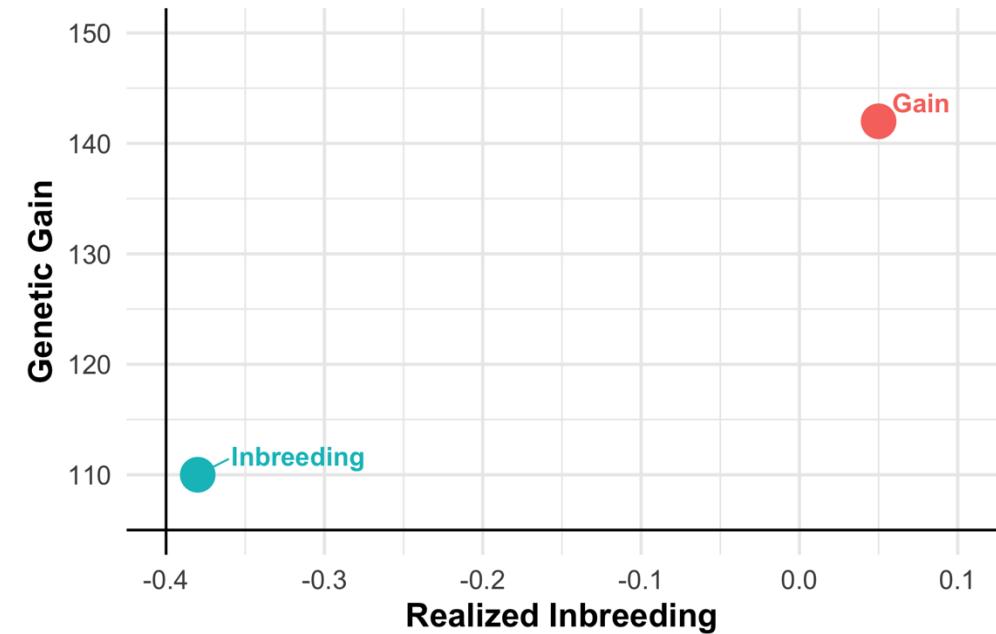
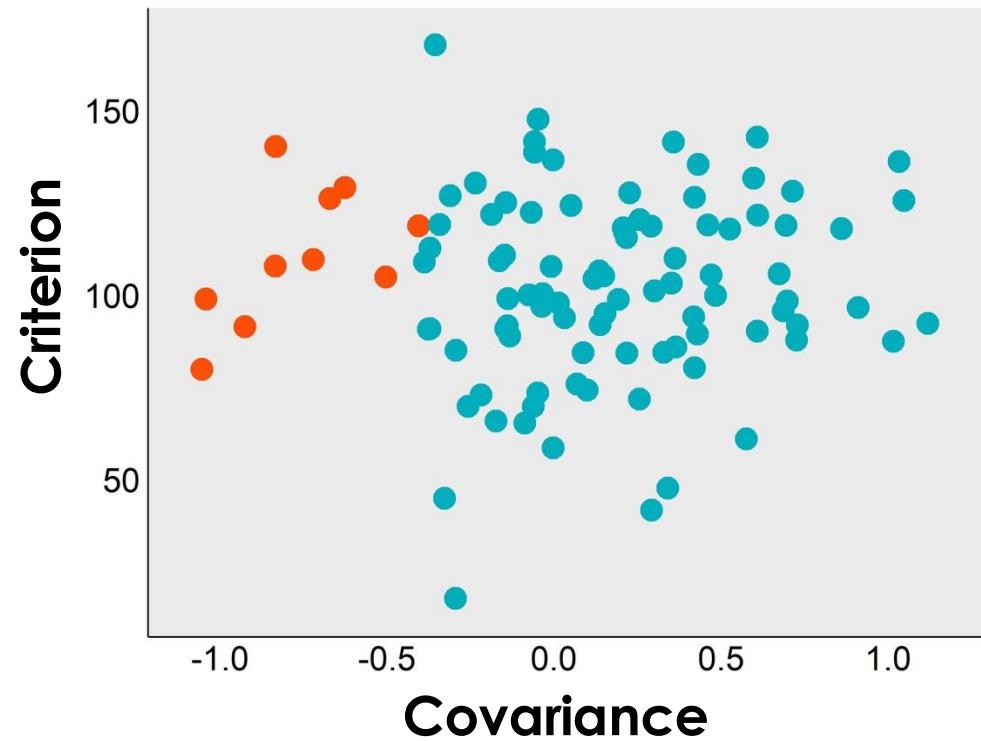
Optimal cross selection



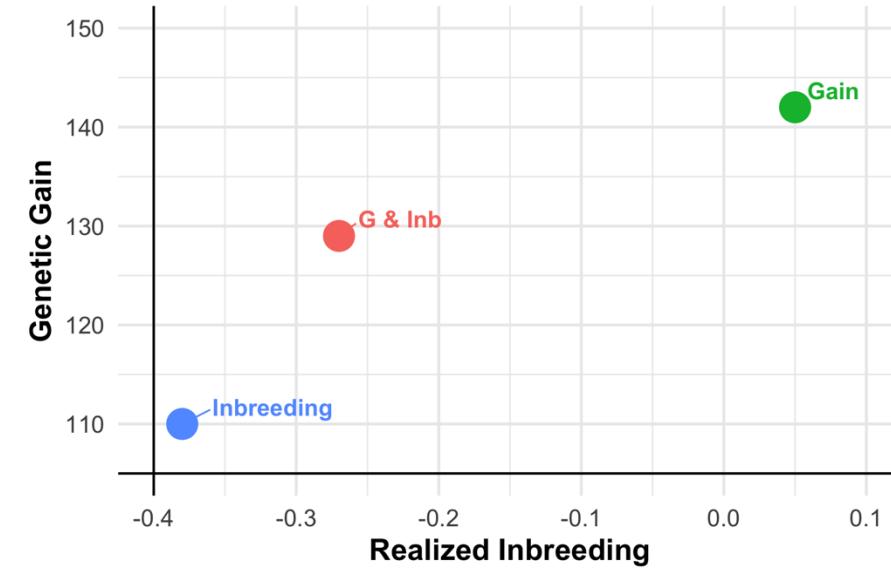
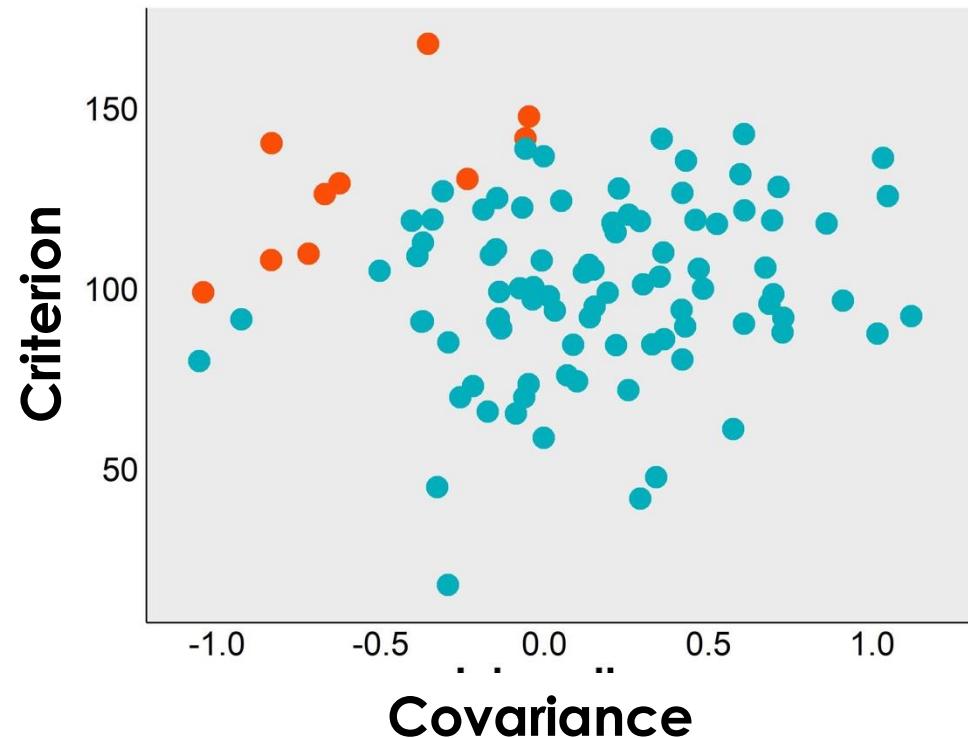
$$\frac{A}{2} = \text{Coancestry}$$

Coancestry (t) = Inbreeding ($t+1$)

Optimal cross selection



Optimal cross selection



- Cross prediction and optimization

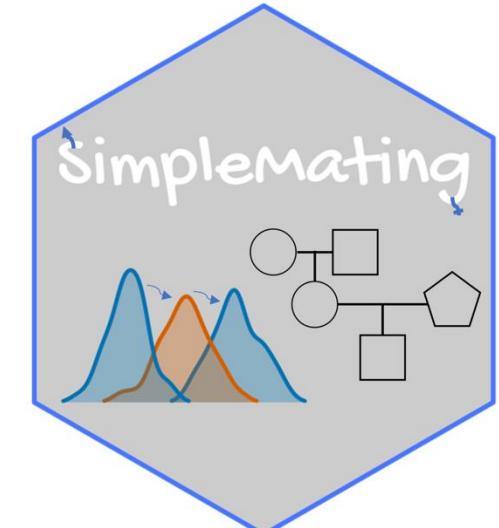
SimpleMating: R-package for prediction and optimization of breeding crosses using genomic selection

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Luís Felipe V. Ferrão³  | Patrício R. Munoz³  | Márcio F. R. Resende Jr.² 

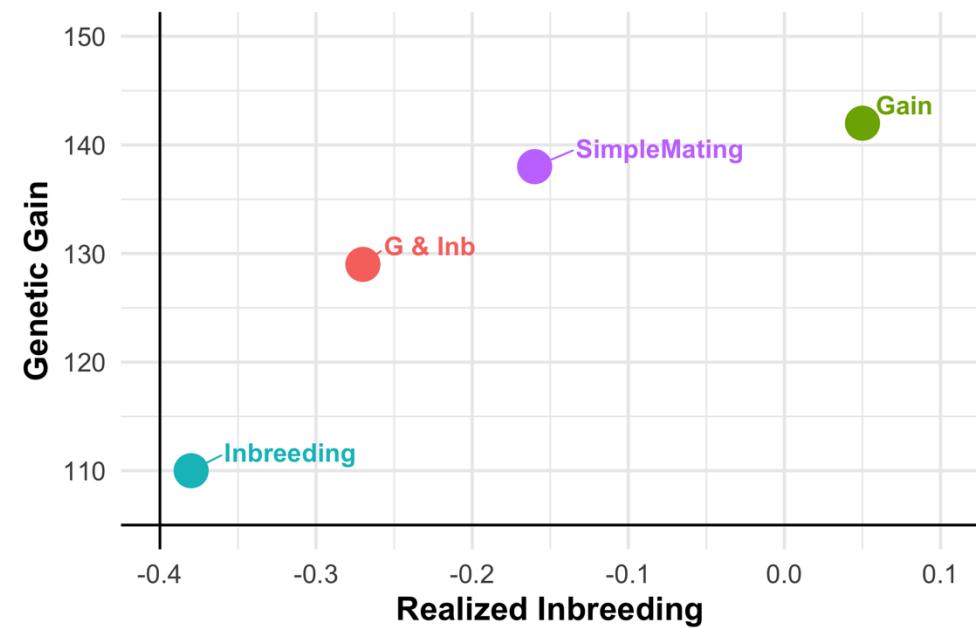
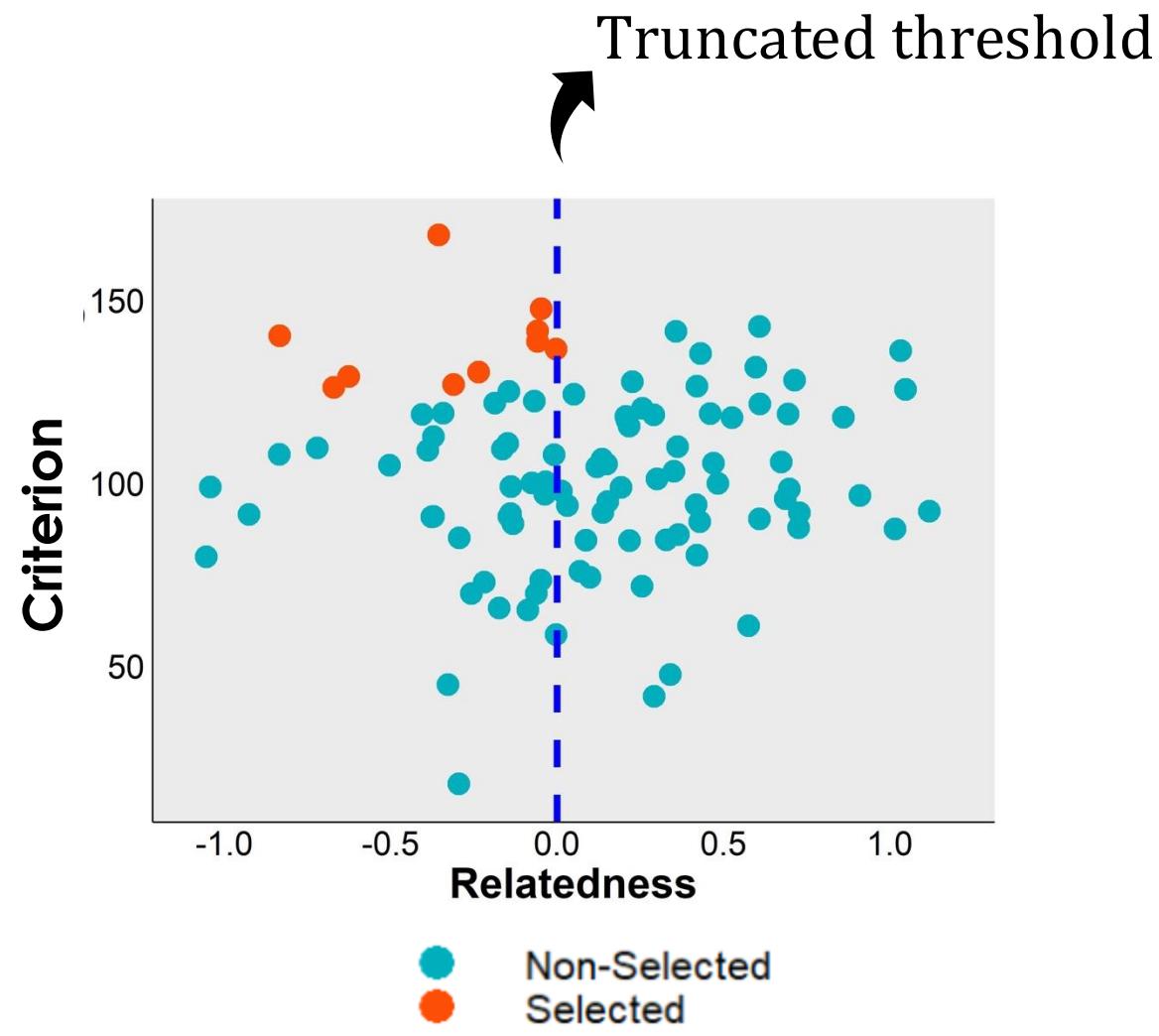
- Two module package

- i. Prediction of cross performance

- ii. Optimization of crosses



Optimal cross selection



SimpleMating: cross performance

1

Mid-parental value

$$\mu_{ebv} = \frac{ebv_{p1} + ebv_{p2}}{2}$$

3

Usefulness (A/AD)

$$UC = \mu + i\sigma$$

2

Total genetic value

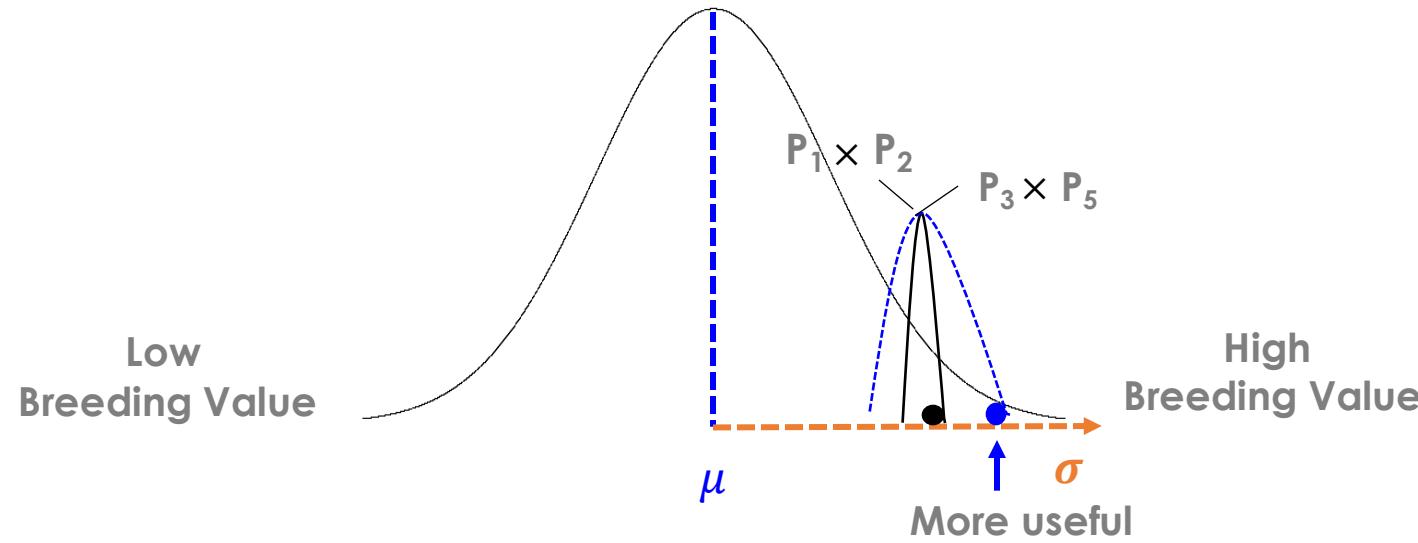
$$\mu_{gv} = \sum_{i=1}^n [a_i(p_i - q_i - y_i) + d_i[2p_iq_i + y_i(p_i - q_i)]]$$

4

Multi-trait for all above

SimpleMating – Cross performance

Prediction of cross performance: Usefulness



$$UC = \mu + i\sigma$$

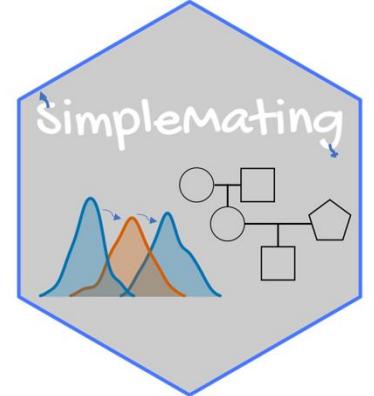
μ = progeny mean

i = selection intensity

σ = progeny standard deviation

Schnell and Utz (1975)

Prediction of cross performance



Output

```
### ----- 1. Performance prediction
ST_tgv2 <- getTGV(MatePlan = CrossPlan,
                     Markers = Markers,
                     addEff = addEff,
                     domEff = domEff,
                     K = relMat,
                     ploidy = 2)
```

```
### ----- 1. Performance prediction
usefAdd <- getUsefA(MatePlan = CrossPlan,
                      Markers = Markers,
                      addEff = effAdd,
                      K = relMat,
                      Map.In = Map, # Genetic map
                      linkDes = NULL,
                      propSel = 0.05,
                      Type = "DH",
                      Generation = 1)
```

SimpleMating Module 2



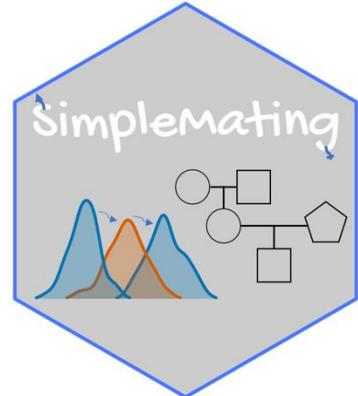
Output

```
### ----- 2. Optimization
MatePlan = selectCrosses(data = ST_tgv4,
                           n.cross = 20,
                           max.cross = 2,
                           min.cross = 1,
                           culling.pairwise.k = 0)
```

```
# stats
MatePlan[[1]]
```

```
# Plan
MatePlan[[2]]
```

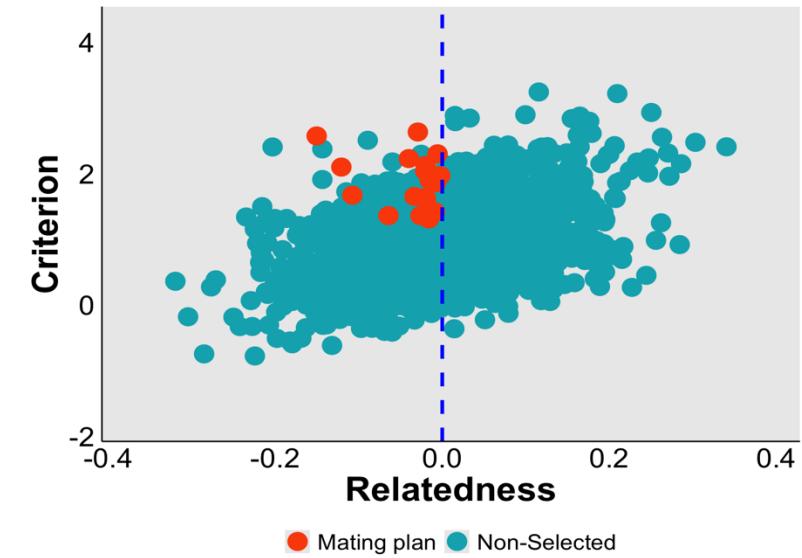
```
# Plot
MatePlan[[3]]
```



Mate plan

P1	P2	Y	K
74	G58	14.55678	-0.11572192
81	G61	12.30788	-0.06813125
79	G61	10.95003	-0.02893493
97	G8	9.72175	-0.01851278
96	G8	9.71596	-0.01758385
28	G23	9.52128	-0.09772107
41	G33	9.36602	-0.01468378
26	G23	9.18741	-0.08279022
32	G27	9.14856	-0.10668185
6	G11	9.02959	-0.06179866

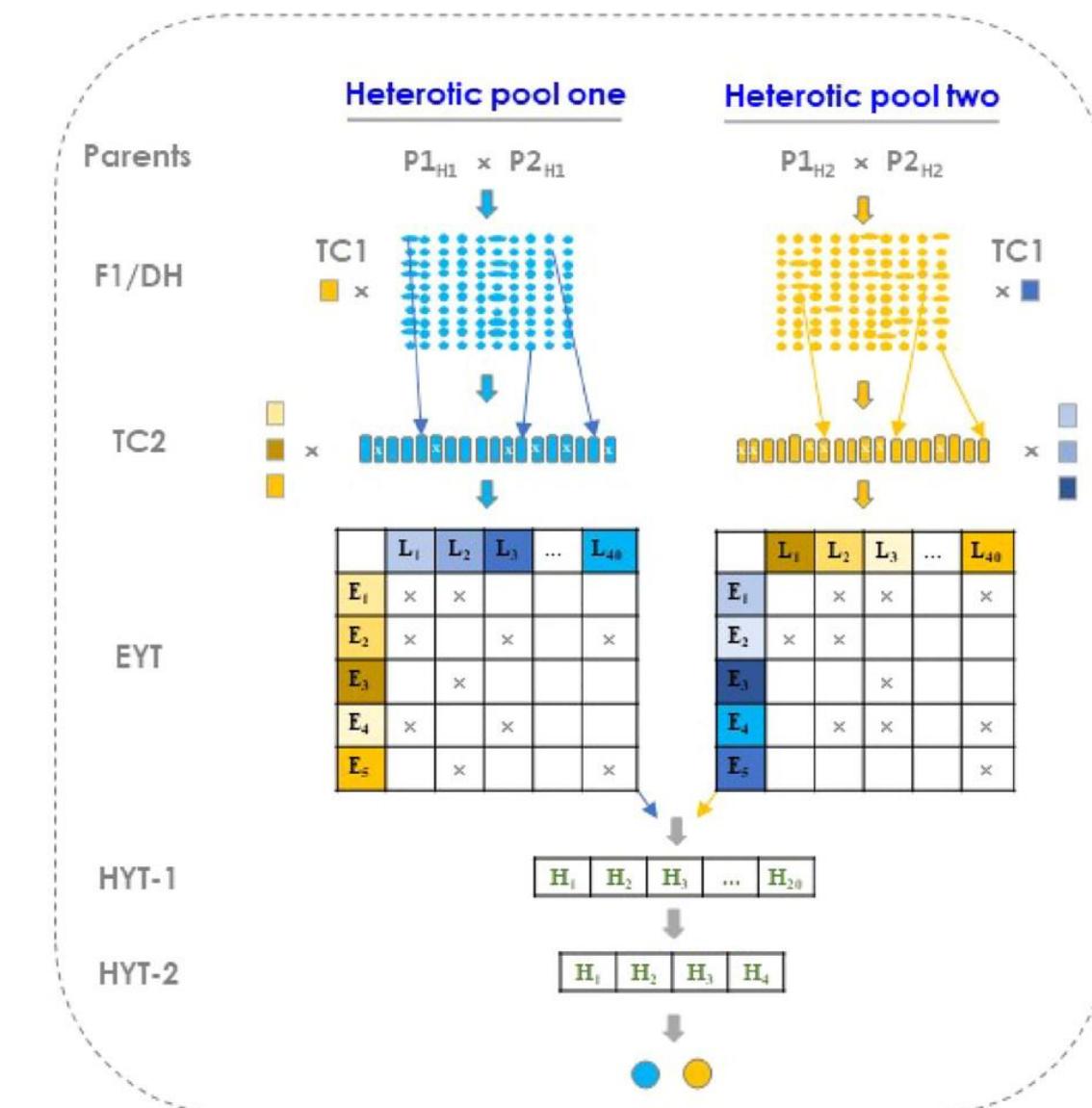
Criteria



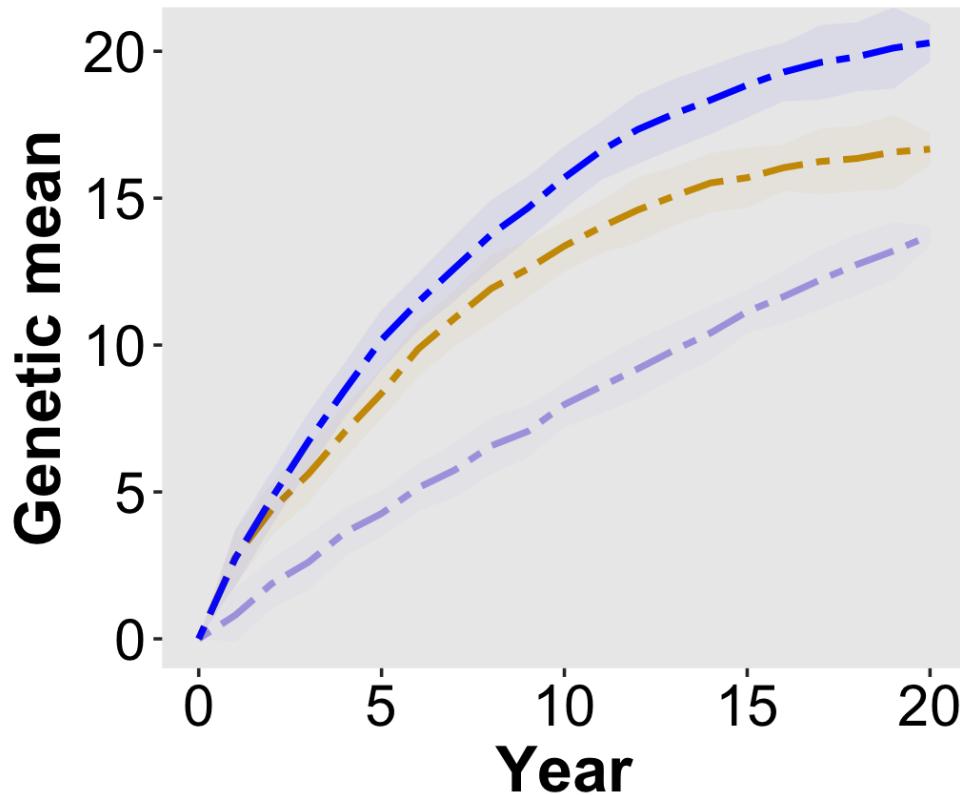
Relationship

Example one: Hybrid breeding program

- Maize breeding program with two heterotic groups;
- Recurrent reciprocal selection scheme;
- PS: selection only using phenotypes;
- GS: Truncated selection via estimated breeding values;
- OCS: Optimal cross selection with usefulness as criterion (SimpleMating);
- 20 years of simulations (AlphasimR).

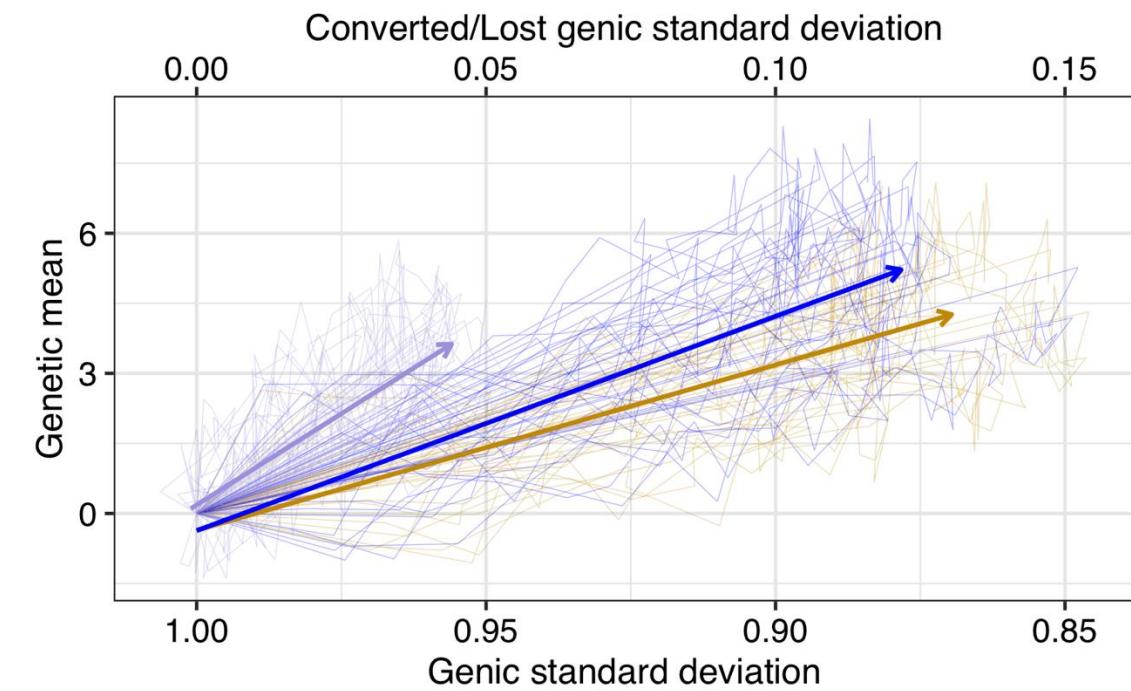
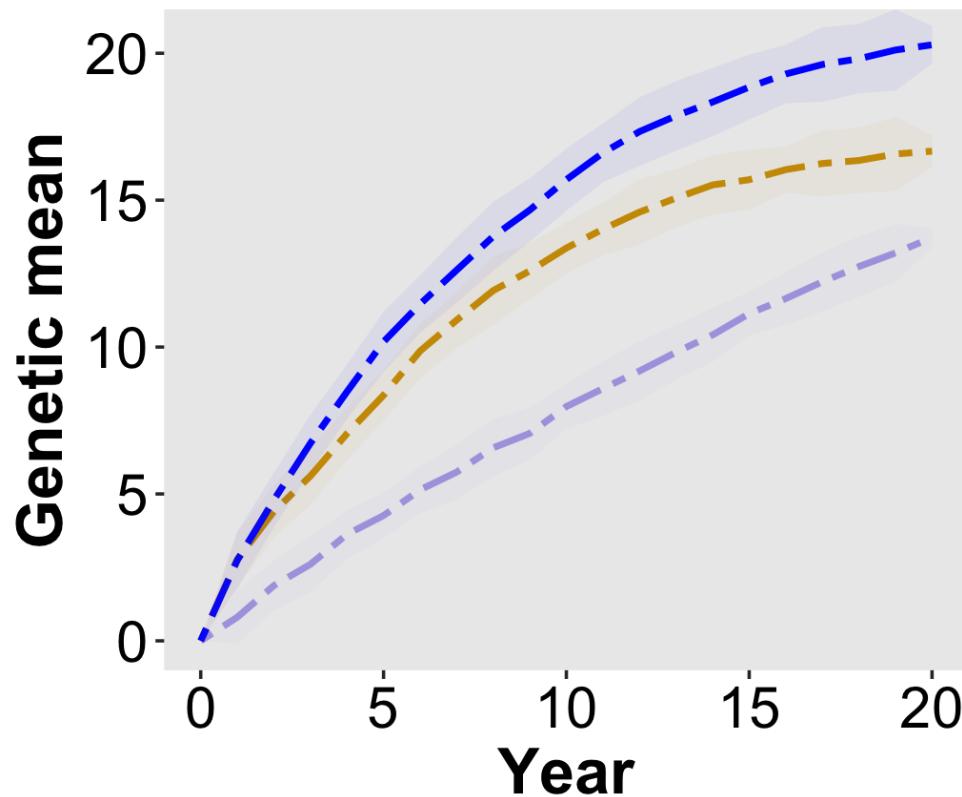


Example one: Hybrid breeding program



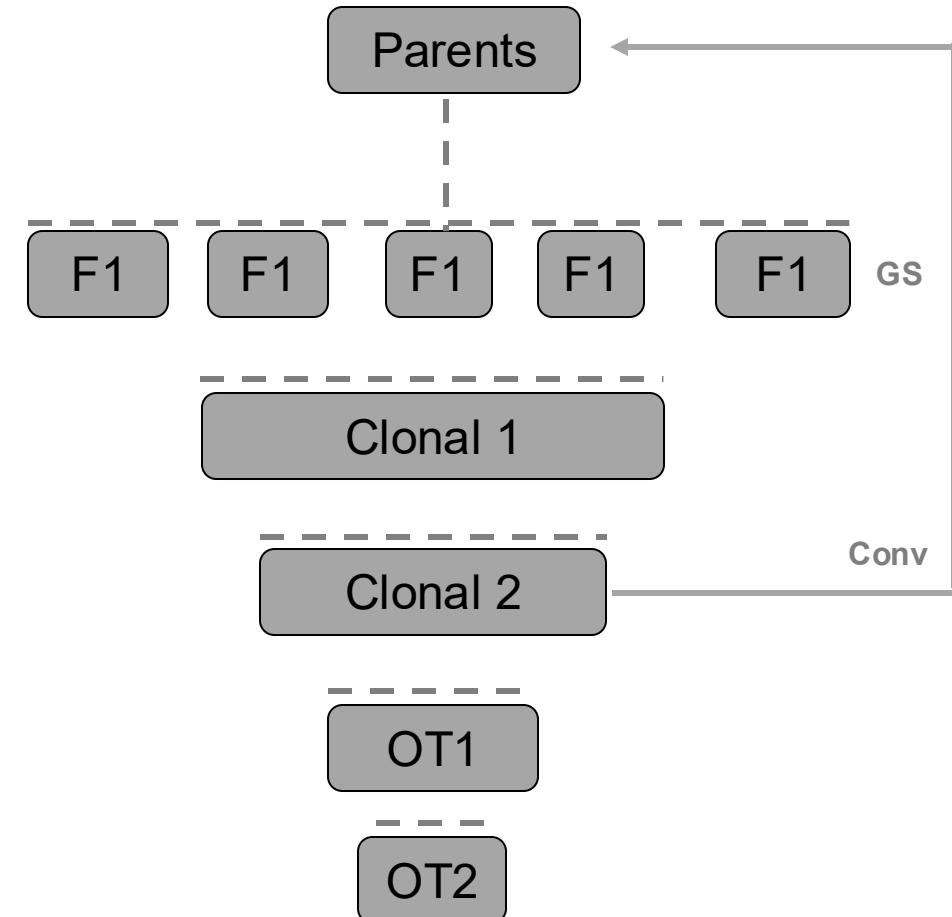
→ GS → OCS → PS

Example one: Hybrid breeding program

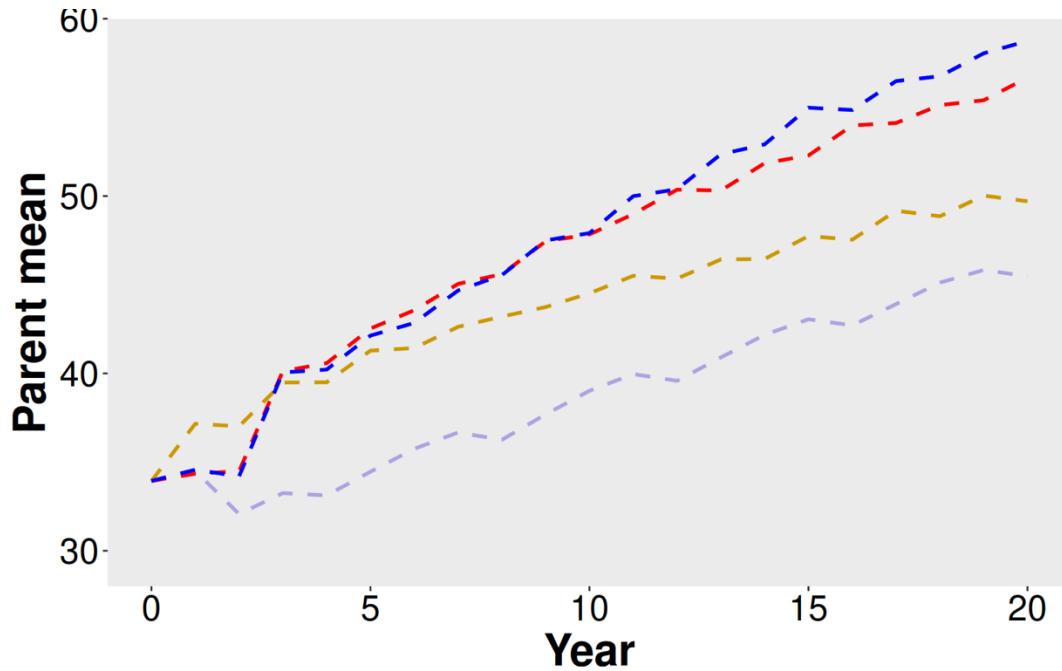


Example Two: Clonal breeding program

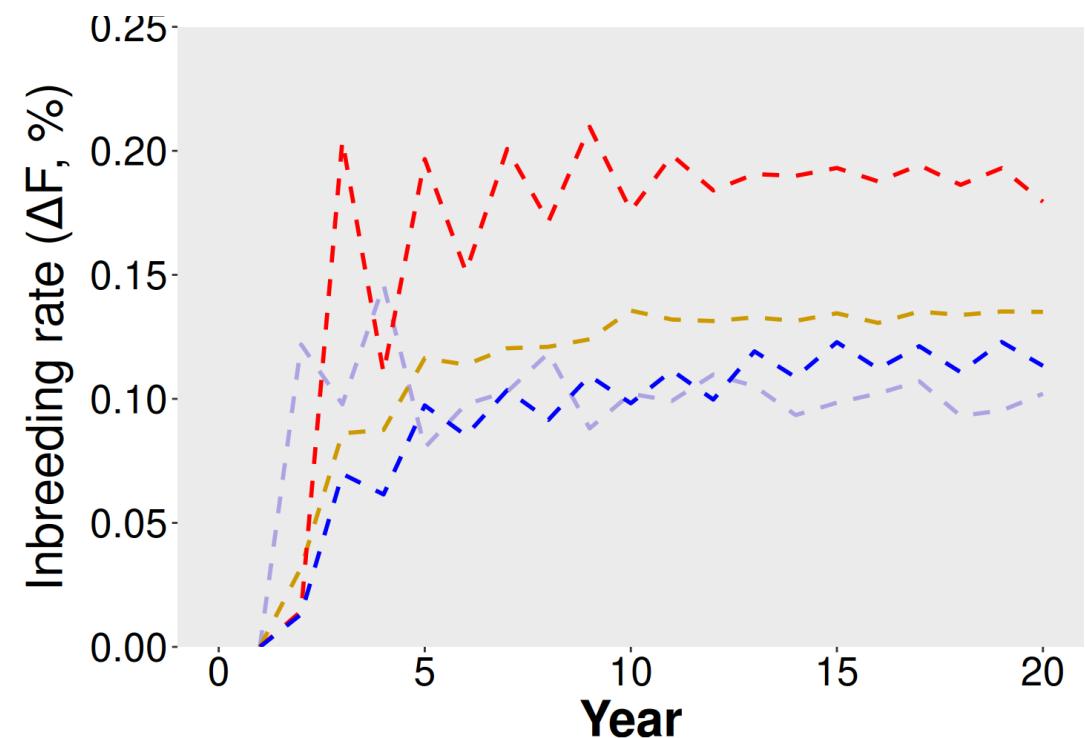
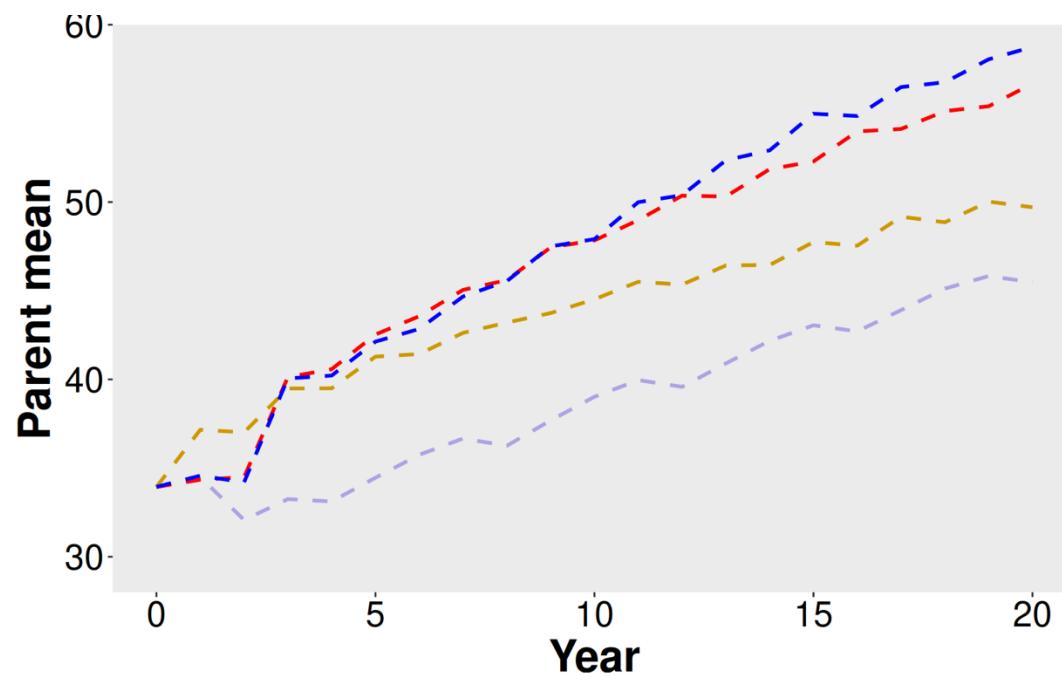
- **Clonal** breeding program (potato UF);
- **Recurrent reciprocal selection** scheme;
- **PS**: selection only using phenotypes;
- **GS**: Truncated selection via estimated breeding values;
- **TGV**: Truncated selection via total genetic value (A+D);
- **OCS**: Optimal cross selection with TGV as criterion (SimpleMating);
- **20 years of simulations** (AlphasimR);
- **Dominance deviation**: 0.4.



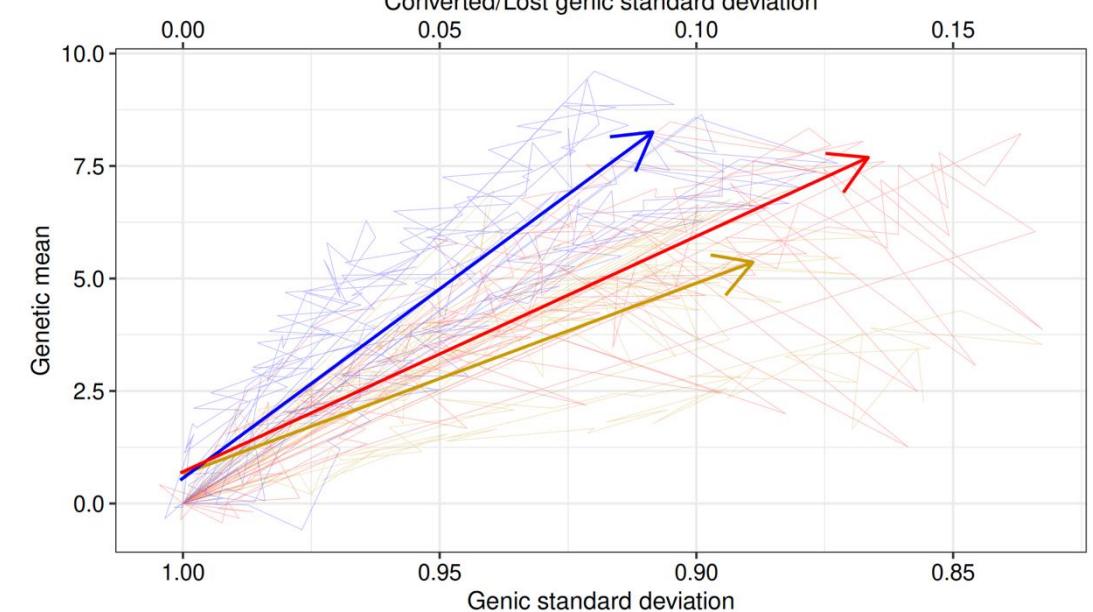
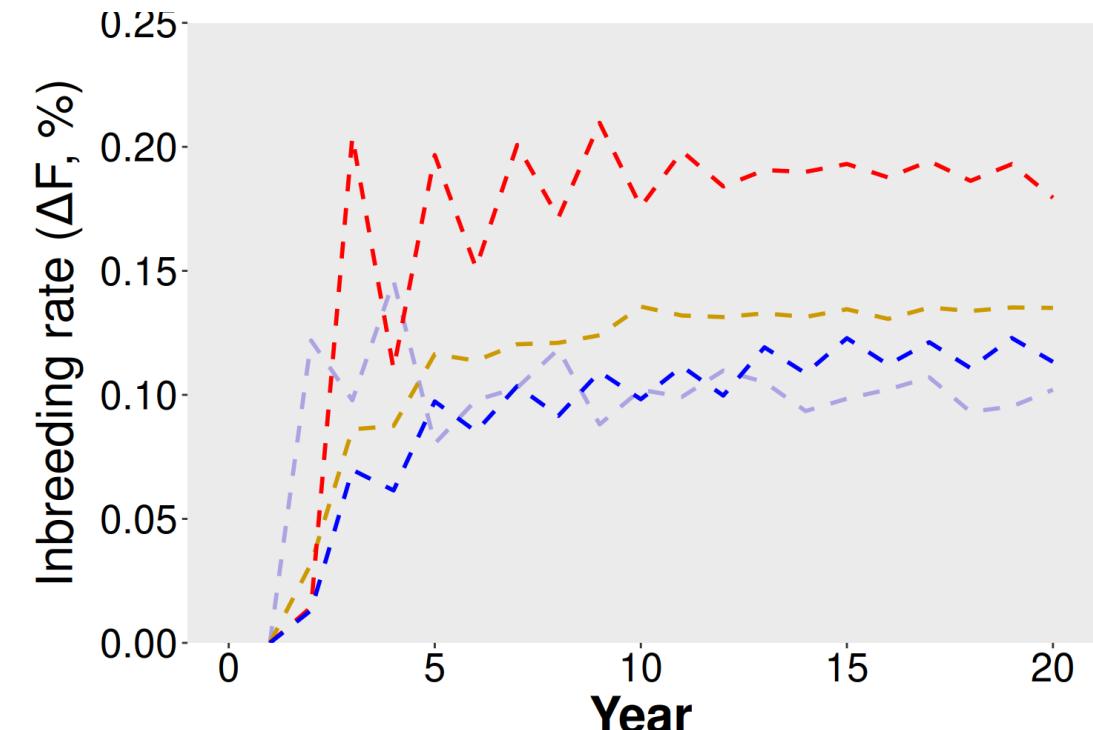
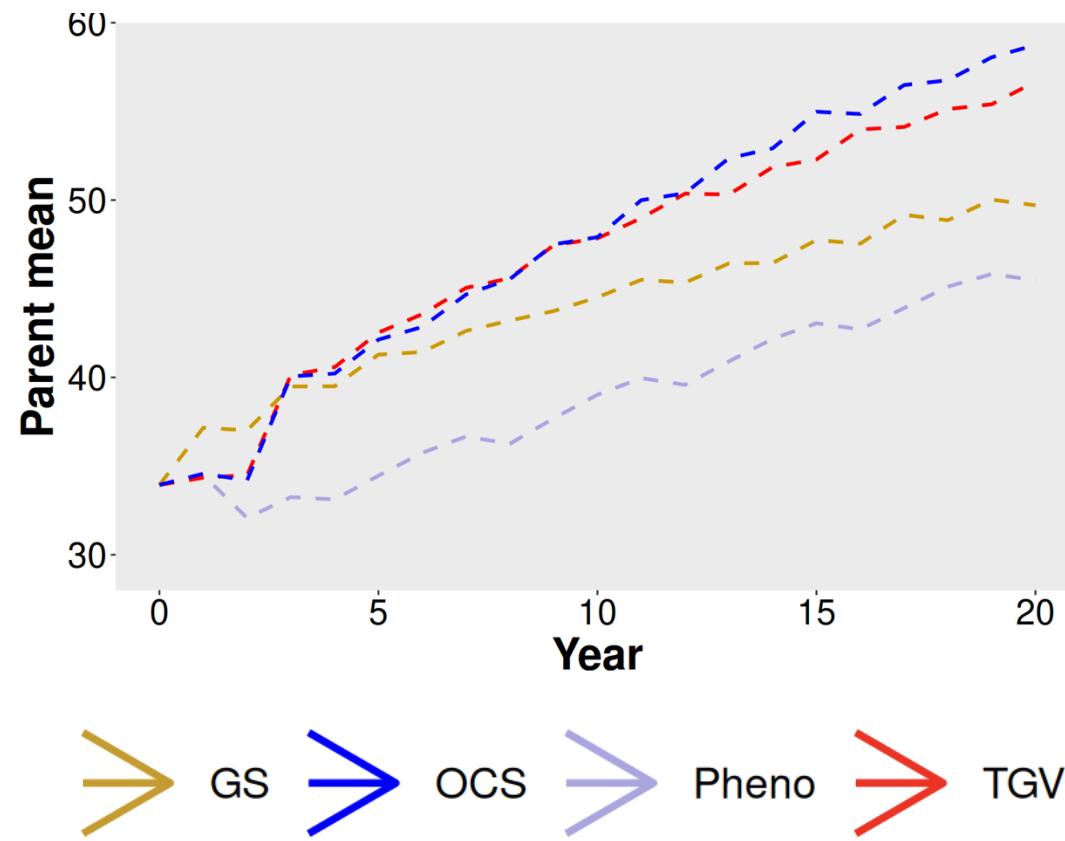
Example Two: Clonal breeding program



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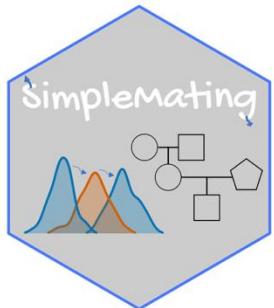


Example Two: Clonal breeding program





Use of **optimal cross selection** is an efficient implementation in various genomic selection-based breeding programs



SimpleMating provides an efficient framework for optimizing crosses and improving selection decisions in breeding programs

Acknowledgments

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<https://marcopxt.github.io/>

